

# Hackathon1

October 10, 2023

## 1 Hackathon Stat

This project illustrates the course LEPL1109 with an industrial application of statistics. You will analyse the capacity of solar production of electricity located in the French cities of Caen and Tours. The file ‘radiation.csv’ contains 3 columns DATE : YYYYMMDD, Caen and Tours : the daily solar radiation in W/m2 measured in the 2 cities. Notice that data for some days are not reported due to failure of measurement system.

### 1.1 Report content

- You have to fill in this jupyter notebook downloadable on the moodle website of the course
- Grades are granted to the members whose names are in the Jupyter notebook. If your name doesn’t appear on the top of the notebook, you’ll get a 0, even though you are in a group on Moodle.
- The jupyter notebook must be compiled with printed results and next submitted via moodle. The absence of compiled results (or non-printed values) leads to a lower grade.

### 1.2 Report submission

- The deadline for submission is reported on the moodle website. Submission after the deadline will not be accepted.
- To submit your report, go to the section “APP” on Moodle and the subsection “Soumission du rapport”. You can upload your work there. Once you are sure that it is your final version, click the button “Envoyer le devoir”. It is important that you don’t forget to click on this button !
- Reports that have not been uploaded through Moodle will not be corrected.

### 1.3 Names and Noma of participants:

Part. 1: Arthur De Neyer - 74842100

Part. 2: Axel Snessens - 29202100

Part. 3: Andrea Dalmasso - 16012100

Part. 4: Meriemque Marie - 59971900

Part. 5: Nouha El Hariri - 03032100

Part. 6: Aylin Defoy - 62322100

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## 1.4 1. Energy calculation and basic statistics

Compute the daily energy in WH per square meter of solar panel. For this purpose you use the datasets reporting the solar irradiation measure in Caen and Tours (source <https://www.ecad.eu/>). The irradiation is measured in W/m2 per day. You will use the formula:

$$C = E_{\text{Sol}} \times 24 \times P_{\text{cr}} \times f_{\text{perf}}$$

where

C is the electricity produced in WH/m2 for a day

E\_sol is the daily solar radiation in W/m2

P\_cr is the peak power coefficient, set here to 0.18 (monocrystal silicium)

f\_perf depends upon the system, set here to 0.75.

Remark:

$$1 \text{ W} = 1 \text{ J/sec}$$

$$1 \text{ WH is } 1 \text{ W} \times 3600 \text{ sec} = 3600 \text{ J}$$

$$\text{energy/m2} = E_{\text{sol}} \times 24 \times 3600 \text{ J/m2} = E_{\text{sol}} \times 24 \text{ WH/m2}$$

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1.1. Start by computing the daily energy in WH produced by a 1m2 solar panel

- Plot time-series of solar electric production in Caen and Tours from 1974 to 2023. Comment the evolution.
- Plot boxplots of daily productions for both cities. Comment the box plot.
- Remove outliers using the interquartile range.
- Plot an histogram of daily electricity production, after removal of outliers.

Watchout: remove all days for which a outlier is observed in Caen **or** Tours to keep the same number of observations.

```
[ ]: import pandas as pd
import matplotlib.pyplot as plt
import matplotlib.dates as mdates

# ===== Load CSV & convert dates =====

data = pd.read_csv('Radiation.csv')

data['DATE'] = pd.to_datetime(data['DATE'], format='%Y%m%d')

# parsing data to keep 1 line out of 100 (better visibility, less precision)
data = data.iloc[range(0, len(data), 100), :]
```

```

# ===== Plot settings =====

# taille
plt.figure(figsize=(36, 12))
plt.xlim(pd.Timestamp('1974-01-01'), pd.Timestamp('2023-12-31'))

# étiquette arrière plan
plt.grid(True)
years = mdates.YearLocator()
year_fmt = mdates.DateFormatter('%Y')
plt.gca().xaxis.set_major_locator(years)
plt.gca().xaxis.set_major_formatter(year_fmt)
plt.xticks(rotation=45)

# légende
plt.xlabel('Date')
plt.ylabel('solar electric production (WH/m²)')
plt.title('solar electric production in Caen and Tours from 1974 to 2023')
plt.legend()

# ===== Plot =====

# Apply formula:  $C = E_{Sol} \times 24 \times P_{cr} \times f_{perf}$  for both cities data
plt.plot(data['DATE'], data['Caen']*24*0.18*0.75, label='Caen',
        color='darkgoldenrod')

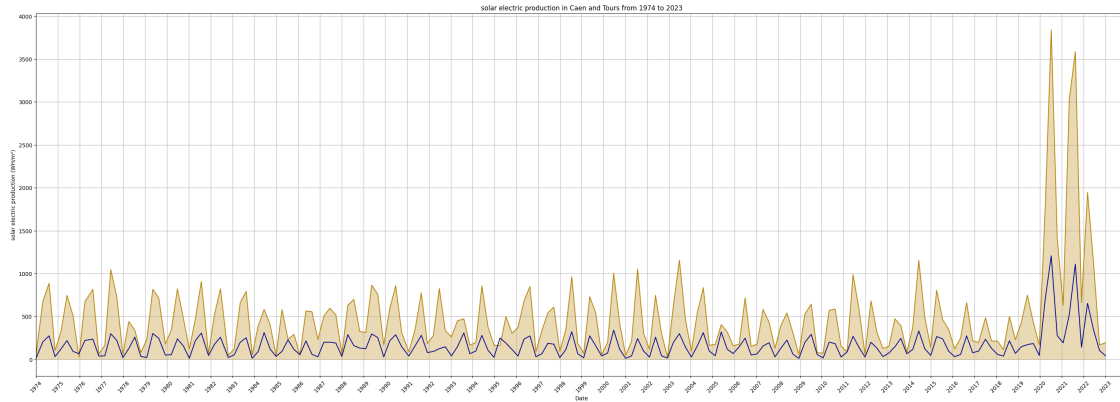
plt.fill_between(data['DATE'], data['Caen']*24*0.18*0.75,
        color='darkgoldenrod', alpha=0.3)

plt.plot(data['DATE'], data['Tours'], label='Tours', color='darkblue')

# ===== Show Plot =====
plt.show()

```

No artists with labels found to put in legend. Note that artists whose label start with an underscore are ignored when legend() is called with no argument.




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1.2. We want to compute monthly statistics of electricity solar production. Calculate for each city and for each month:

- 1) the average daily production of electricity in Wh/m2
- 2) the median daily production of electricity in Wh/m2
- 3) the standard deviation daily production of electricity in Wh/m2
- 4) the 5% percentile of daily production of electricity in Wh/m2
- 5) the 95% percentile of daily production of electricity in Wh/m2

Report the results in one or two tables.

Compare and comment these statistics!

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## 1.5 2. Fit of distributions and hypothesis tests

2.1. We focus on the daily production of electricity in April. Retrieve the data for month of April, in Caen and Tours.

- 1) Fit Gamma and normal distributions by log-likelihood maximization to daily production of electricity during April (Caen & Tours).
- 2) Compute the 4 log-likelihoods and select the best model for each location (justify your answer).
- 3) Compare on the same plot the empirical, the gamma and normal pdf (the empirical pdf is an histogram of frequencies).
- 4) Why is there 3 parameters in python for the Gamma pdf whereas there is only 2 in the distribution seen during lectures?

Remark : set `floc` to -0.001 for the `gamma.fit` (to avoid troubles in case of null observations)

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2.2. Check if the average daily production in April is the same in Caen and Tours. Let us recall that the null hypothesis is

$$H_0: \mu_{Caen} = \mu_{Tours}.$$

Take care to comment your conclusions. Are all assumptions required to perform this test satisfied?

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2.3. Test the equality of variance of daily production in April at Caen & Tours?  $H_0: \sigma_{Caen} = \sigma_{Tours}$ .

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2.4. Explain the Wilcoxon's test. What is the main advantage of this test compared to the Student's T test. Why is this useful in our project?

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2.5. Apply the Wilcoxon test to distributions of daily productions in April, at Caen and Tours. What can you conclude about the means of daily production in these 2 cities?

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## 1.6 3. Regression and forecasting

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3.1. Do we observe any trend in the yearly solar production of electricity over the considered period? To answer this question:

- You will compute the average daily production (Wh/m<sup>2</sup>) during April from 1977 up to 2019 (included).
- You get a time-series of 44 values for each city. Regress these values on the explanatory variables  $X=(\text{Year}-1977)$ . Don't forget to add a constant term and analyze results.
- Plot on the same graph, the predicted and the observed values.
- Comment your results!

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3.2. You want to design a model to forecast the solar electric production for the next day (location Caen only). You will work with data over the period 1977 to 2019.

Let us denote by  $C(t)$  the production on day 't'. The model that we want to fit is called autoregressive and is defined as follows:

$$C(t) = \sum_{k=1}^{10} a_k C(t-k)$$

This model is common in time-series analysis and predicts the production of the next day with the recent observations.

- Split the dataset into a training set (1977 to 2010 included) and a validation set (2011 to 2019 included).
- Estimate this model with statsmodels on the training set.
- How would you judge the quality of the predictive model? (Analyze statistics reported by statsmodel)
- Compute the Mean Absolute Error (MAE) between predicted and real consumptions (on the training set).

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3.3. Use this model on the test set to forecast the electric daily production.

- Compare on a graph, the forecast to real consumptions on the given period.
- Plot the errors of prediction. Are they acceptable?
- Compute the MAE on the test set and the  $R^2$ . Is the forecast reliable?

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